

Computer Modeling

Ordinary differential equations
(ODE)

A solution of ODE

An ordinary differential equation of the n -th order

$$G(t, y(t), y'(t), \dots, y^{(n)}(t)) = 0$$

A solution $y(t) = \varphi(t)$ in an interval $t \in (b, d) \subset \mathbb{R}$:

$\varphi(t)$ is n -times differentiable on (b, d) and satisfies

$$G\left(t, \varphi(t), \varphi'(t), \dots, \varphi^{(n)}(t)\right) = 0.$$

$\varphi(t)$ is also called a trajectory.

(we allow $b = -\infty$ and $d = \infty$)

Uniqueness of a solution

A solution is unique, if for any point $(t^*, y^*(t^*))$ there is only one curve $y(t)$ going through this point.

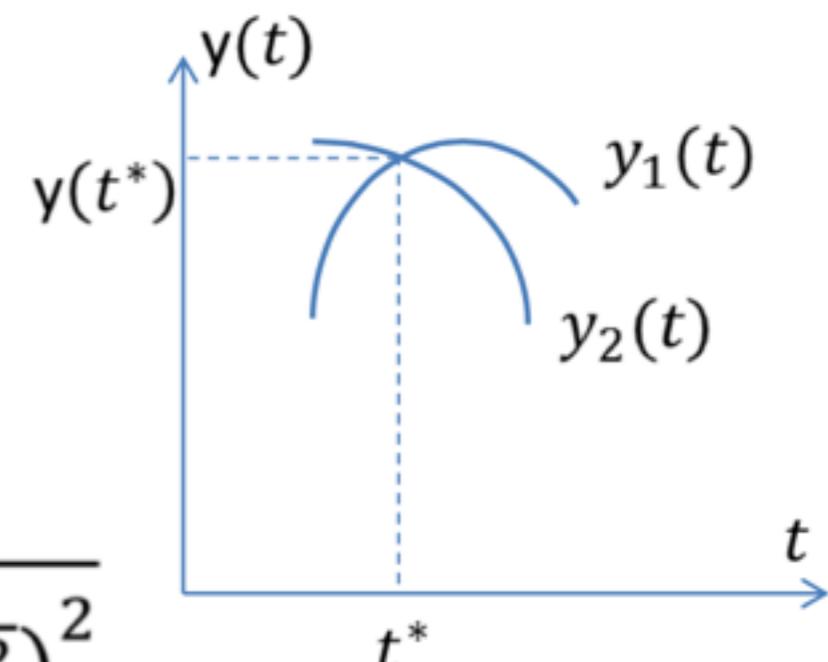
A solution may be non-unique,
e.g. for the following equation

$$y(t)\sqrt{1-t^2}y'(t) = t\sqrt{1-y^2(t)}$$

two solutions go through the point $(0,1)$:

$$y_1(t) = 1 \text{ and } y_2(t) = \sqrt{1 - (1 - \sqrt{1-t^2})^2}$$

$$y'_1(t) = 0 \text{ and } y'_2(t) = \frac{t(1-\sqrt{1-t^2})}{\sqrt{1-t^2}\sqrt{1-(1-\sqrt{1-t^2})^2}} = \frac{t(1-\sqrt{1-t^2})}{\sqrt{1-t^2}y_2(t)}$$



Types of solutions

A solution may depend on an arbitrary constant, e.g. an equation $y'(t) = h(t)$ has a solution $y(t) = \int h(t)dt + C$.

Consider a domain D containing only points $(t, y) \in D$ such that the solution $y(t)$ going through any point of D is unique. A function $y(t) = \varphi(t, C)$ is called **a general solution** of the 1st order equation, if:

- (i) φ is continuously differentiable with respect to t ,
- (ii) it is possible to find C for any points $(t, y) \in D$,
- (iii) $\varphi(t, C)$ is a solution for any such determined constant.

A general solution of an n -th order equation depends on n arbitrary constants C_1, \dots, C_n .

A particular solution goes through a particular point (t, y) and can be obtained by setting particular values to the constants.

A singular solution is nonunique in all points (t, y) of its trajectory.

An example

$$y'(t) = 2\sqrt{y(t)}, \quad y \geq 0$$

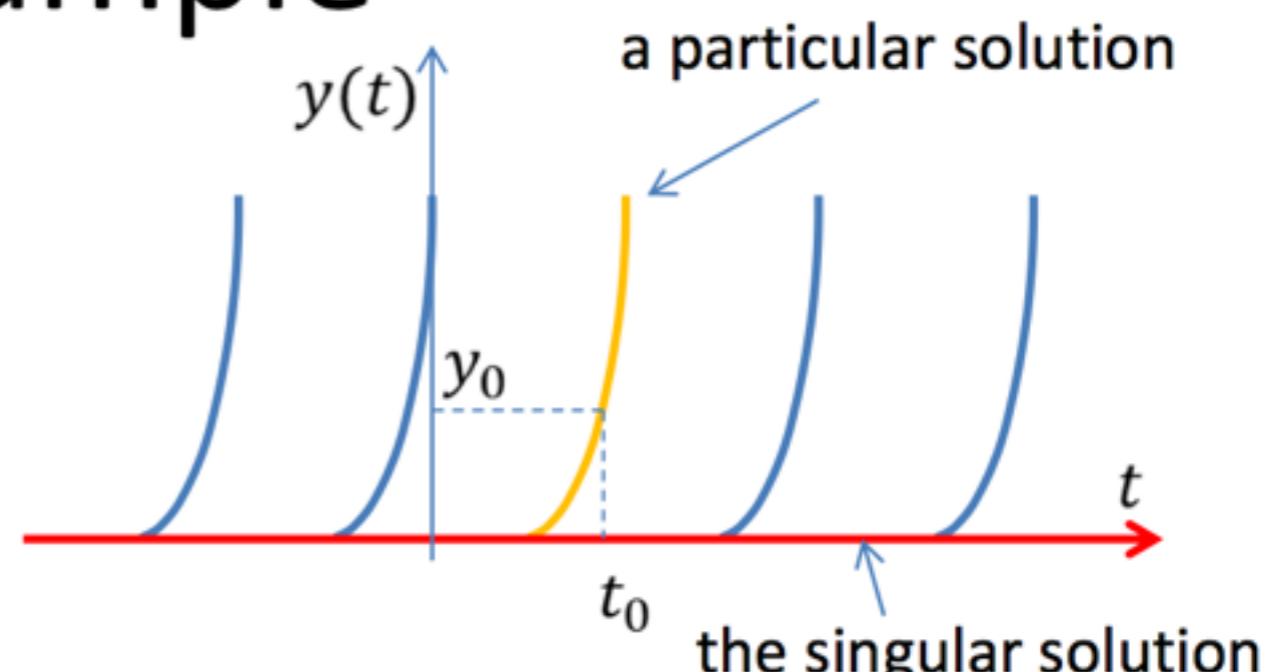
$y(t) = 0$ is a solution

assume $y(t) > 0$ for any t

$$\frac{y'}{2\sqrt{y}} = 1 \rightarrow (\sqrt{y})' = 1$$
$$\sqrt{y} = t + C, \quad C > -t$$

all solutions

$$y(t) = (t + C)^2, \quad C \geq -t$$



$$D = \{(t, y) : -\infty < t < \infty, y > 0\}$$

A general solution:

$$y(t) = (t + C)^2, \quad t > -C$$

A particular solution going through (t_0, y_0) :

$$y(t) = (t - t_0 + \sqrt{y_0})^2$$

The singular solution $y(t) = 0$
(nonunique at all points of its trajectory)

Initial and boundary conditions

Initial value problem:

conditions for operators $A_i(t, y) = 0$, $i = 1, \dots, n$
at a t_0 such that $(t_0, y(t_0)) \in D$

Cauchy problem:

$$y(t_0) = y_0^{(0)}, y'(t_0) = y_1^{(0)}, \dots, \\ y^{(n-1)}(t_0) = y_{n-1}^{(0)}, \quad (t_0, y(t_0)) \in D$$

Boundary value problem: conditions distributed among two different points t_0 and t_1

System of ODE equations

$$G(t, y(t), y'(t), \dots, y^{(m)}(t)) = 0$$

Denote

$$y_1(t) = y(t)$$

$$y_2(t) = y'(t)$$

.....

$$y_m(t) = y^{(m-1)}(t)$$

then

$$y'_1(t) = y_2(t)$$

$$y'_2(t) = y_3(t)$$

.....

$$y'_{m-1}(t) = y_m(t)$$

$$G(t, y_1(t), y_2(t), \dots, y'_m(t)) = 0$$

If the equation

$$G(t, y_1(t), y_2(t), \dots, y'_m(t)) = 0$$

can be solved for $y'_m(t)$, then the equation can be written in the normal form. In general

$$y'_1(t) = f_1(t, y_1, y_2, \dots, y_M)$$

$$y'_2(t) = f_2(t, y_1, y_2, \dots, y_M)$$

.....

$$y'_{M-1}(t) = f_{M-1}(t, y_1, y_2, \dots, y_M)$$

$$y'_M(t) = f_M(t, y_1, y_2, \dots, y_M)$$

System of linear equations

System of linear equations:

$$y'_1(t) = a_{11}(t)y_1(t) + \cdots + a_{1M}(t)y_M(t) + u_1(t)$$

$$y'_2(t) = a_{21}(t)y_1(t) + \cdots + a_{2M}(t)y_M(t) + u_2(t)$$

.....

$$y'_M(t) = a_{1M}(t)y_1(t) + \cdots + a_{MM}(t)y_M(t) + u_M(t)$$

Picard theorem. If $a_{ij}(t), i, j = 1, 2, \dots, M$ and $u_i(t), i = 1, 2, \dots, M$ are continuous in an interval $[b, d]$, then the above system has the unique solution $y_1(t), y_2(t), \dots, y_M(t)$, which satisfies initial conditions $y_1^{(0)}, y_2^{(0)}, \dots, y_M^{(0)}$ at any $t \in [b, d]$ for arbitrary values $y_1^{(0)}, y_2^{(0)}, \dots, y_M^{(0)}$. The solution exists and is continuously differentiable in (b, d) . \rightarrow no singular solutions

Homogeneous and nonhomogeneous systems

The system of equations is called **homogeneous**, if
 $u_1(t) = u_2(t) = \dots = u_M(t) \equiv 0$

Otherwise it is called **nonhomogeneous**.

Homogeneous system

$$\begin{aligned}y'_1(t) &= a_{11}(t)y_1(t) + \cdots + a_{1M}(t)y_M(t) \\y'_2(t) &= a_{21}(t)y_1(t) + \cdots + a_{2M}(t)y_M(t) \\&\dots \dots \dots \dots \dots \dots \quad (*)\end{aligned}$$

$$y'_M(t) = a_{1M}(t)y_1(t) + \cdots + a_{MM}(t)y_M(t)$$

Assumption: $a_{ij}(t)$ are continuous in $[b, d]$

If any column of functions ($1 \leq m \leq M$)

$$y_{11}(t), \dots, y_{m1}(t)$$

$\dots \dots \dots \dots \dots \dots$

$$y_{1M}(t), \dots, y_{mM}(t)$$

is a solution of (*), then their linear combination

$$y_1(t) = h_1 y_{11}(t) + \dots + h_m y_{m1}(t)$$

$\dots \dots \dots \dots \dots \dots$

$$y_M(t) = h_1 y_{1M}(t) + \dots + h_m y_{mM}(t)$$

is also a solution (proof by insertion to (*))

Sketch of a proof ($M = 2$)

$$y'_1(t) = a_{11}(t)y_1(t) + a_{12}(t)y_2(t)$$

$$y'_2(t) = a_{21}(t)y_1(t) + a_{22}(t)y_2(t)$$

Solutions

$$y_{11}(t), y_{21}(t)$$

$$y_{12}(t), y_{22}(t) \quad \text{i.e.}$$

$$y'_{11}(t) = a_{11}(t)y_{11}(t) + a_{12}(t)y_{12}(t)$$

$$y'_{12}(t) = a_{21}(t)y_{11}(t) + a_{22}(t)y_{12}(t)$$

$$y'_{21}(t) = a_{11}(t)y_{21}(t) + a_{12}(t)y_{22}(t)$$

$$y'_{22}(t) = a_{21}(t)y_{21}(t) + a_{22}(t)y_{22}(t)$$

Linear combinations

$$\tilde{y}_1(t) = h_1 y_{11}(t) + h_2 y_{21}(t)$$

$$\tilde{y}_2(t) = h_1 y_{12}(t) + h_2 y_{22}(t)$$

Differentiating the first equation

$$\tilde{y}'_1(t) = h_1 y'_{11}(t) + h_2 y'_{21}(t) =$$

$$= h_1 [a_{11}(t)y_{11}(t) + a_{12}(t)y_{12}(t)] +$$

$$+ h_2 [a_{11}(t)y_{21}(t) + a_{12}(t)y_{22}(t)] =$$

$$= a_{11}(t)[h_1 y_{11}(t) + h_2 y_{21}(t)] +$$

$$+ a_{12}(t)[h_1 y_{12}(t) + h_2 y_{22}(t)] =$$

$$= a_{11}(t)\tilde{y}_1 + a_{12}(t)\tilde{y}_2$$

Similarly for the second equation.

Linear dependence

M sets of functions

$$y_{11}(t), \dots, y_{M1}(t)$$

\dots

(**)

$$y_{1M}(t), \dots, y_{MM}(t)$$

are called **linearly dependent** in an interval (b, d) , if and only if there exists a set of numbers h_1, \dots, h_M , not all zero, such that for all $t \in (b, d)$

$$h_1 y_{11}(t) + \dots + h_M y_{1M}(t) = 0$$

\dots

$$h_1 y_{M1}(t) + \dots + h_M y_{MM}(t) = 0$$

If such numbers do not exist, then the sets of functions are **linearly independent**.

Example (f)

$M=3$ sets of functions, with

$\lambda_i \neq \lambda_j$ for $i \neq j$,

$$s_{11}e^{\lambda_1 t}, s_{21}e^{\lambda_1 t}, s_{31}e^{\lambda_1 t}$$

$$s_{12}e^{\lambda_2 t}, s_{22}e^{\lambda_2 t}, s_{32}e^{\lambda_2 t}$$

$$s_{13}e^{\lambda_3 t}, s_{23}e^{\lambda_3 t}, s_{33}e^{\lambda_3 t}$$

and with at least one nonzero s_{ij} in each row, are linearly independent in the interval $(-\infty, \infty)$.

Consider

$$h_1 s_{11} e^{\lambda_1 t} + h_2 s_{12} e^{\lambda_2 t} + h_3 s_{13} e^{\lambda_3 t} = 0$$

$$h_1 s_{21} e^{\lambda_1 t} + h_2 s_{22} e^{\lambda_2 t} + h_3 s_{23} e^{\lambda_3 t} = 0 \quad \text{Similarly for other equations.}$$

$$h_1 s_{31} e^{\lambda_1 t} + h_2 s_{32} e^{\lambda_2 t} + h_3 s_{33} e^{\lambda_3 t} = 0$$

Sketch of a proof

Assume $s_{13} \neq 0$. Consider the first equation

$$h_1 s_{11} e^{\lambda_1 t} + h_2 s_{12} e^{\lambda_2 t} + h_3 s_{13} e^{\lambda_3 t} = 0$$

Multiplying both sides by $e^{-\lambda_1 t}$

$$h_1 s_{11} + h_2 s_{12} e^{(\lambda_2 - \lambda_1)t} + h_3 s_{13} e^{(\lambda_3 - \lambda_1)t} = 0$$

Differentiating

$$h_2 s_{12} (\lambda_2 - \lambda_1) e^{(\lambda_2 - \lambda_1)t}$$

$$+ h_3 s_{13} (\lambda_3 - \lambda_1) e^{(\lambda_3 - \lambda_1)t} = 0$$

Multiplying by $e^{-(\lambda_2 - \lambda_1)t}$ and differentiating

$$h_3 s_{13} (\lambda_3 - \lambda_1) (\lambda_3 - \lambda_2) e^{(\lambda_3 - \lambda_2)t} = 0$$

As $s_{13} \neq 0, \lambda_3 \neq \lambda_1, \lambda_3 \neq \lambda_2$, then $h_3 = 0$.

Valid for any integer $M > 0$.

The Wronskian

Theorem

If the sets of functions $(**)$ are linearly independent, i.e. there do not exist nonzero values of h_1, \dots, h_M satisfying the equations

$$\begin{aligned} h_1 y_{11}(t) + \dots + h_M y_{1M}(t) &= 0 \\ \dots & \\ h_1 y_{M1}(t) + \dots + h_M y_{MM}(t) &= 0 \end{aligned} \tag{\#}$$

and each set is a solution of the homogeneous system of equation $(*)$, then the determinant

$$W(t) = \begin{vmatrix} y_{11}(t) & \cdots & y_{1M}(t) \\ \vdots & \ddots & \vdots \\ y_{M1}(t) & \cdots & y_{MM}(t) \end{vmatrix} \neq 0$$

for any $t \in (b, d)$.

$W(t)$ – the Wronskian (determinant)

Proof

Assume that $(**)$ are linearly independent and there exist $t_0 \in (b, d)$ that $W(t_0) = 0$. Then there exist not all zero values of h_1^0, \dots, h_M^0 satisfying $(\#)$ at $t = t_0$.

Consider the functions

$$y_m(t) = \sum_{i=1}^M h_i^0 y_{im}(t), \quad m = 1, \dots, M.$$

They are solutions of $(*)$ and $y_m(t_0) = 0$. But $y_m(t) \equiv 0$ are also solutions. From the Picard theorem the solutions are unique (zero). Thus $(\#)$ are satisfied with $h_i = h_i^0, i = 1, \dots, M$, not all equal zero. Finally, the sets of functions $(**)$ are dependent, what contradicts the assumption.

Conclusions

If $W(t_0) = 0$ at any $t_0 \in (b, d)$, then $W(t) \equiv 0$ for any $t \in (b, d)$.

If $W(t) \neq 0$ at any $t \in (b, d)$, then it is different from zero in all points of (b, d) .

General solutions

M linearly independent solutions of a homogeneous system of equations

$$y_{11}(t), \dots, y_{M1}(t) \\ \dots \dots \dots \dots \dots \quad (**)$$

$$y_{1M}(t), \dots, y_{MM}(t)$$

are called a **fundamental system** of solutions.

An example

$$\frac{dy_1}{dt} = \frac{t^2}{1+t^2} y_1 + \frac{1}{1+t^2} y_2$$

$$\frac{dy_2}{dt} = \frac{t^2 - 2t}{1+t^2} y_1 + \frac{2t+1}{1+t^2} y_2$$

$$y_{11} = e^t, \quad y_{21} = -1$$

$$y_{12} = e^t, \quad y_{22} = t^2$$

2 sets of
solutions

Fundamental theorem

If $(**)$ is a fundamental system of solutions of a homogeneous system of equations in (b, d) , then

$$y_1(t) = h_1 y_{11}(t) + \dots + h_M y_{1M}(t) \\ \dots \dots \dots \dots \dots$$

$$y_M(t) = h_1 y_{M1}(t) + \dots + h_M y_{MM}(t)$$

is a set of general solutions in the domain
 $b < t < d, -\infty < y_m < \infty,$

$m = 1, \dots, M$, for an arbitrary set of numbers h_1, \dots, h_M .

$$W(t) = \begin{vmatrix} e^t & -1 \\ e^t & t^2 \end{vmatrix} = e^t(t^2 + 1) > 0$$

$$y_1(t) = h_1 e^t - h_2 \quad \text{a general solution}$$
$$y_2(t) = h_1 e^t + h_2 t^2$$

Matrix equation

$$\mathbf{y} = \begin{bmatrix} y_1 \\ \vdots \\ y_M \end{bmatrix} \quad \mathbf{u} = \begin{bmatrix} u_1 \\ \vdots \\ u_M \end{bmatrix}$$

$$\mathbf{A} = \begin{bmatrix} a_{11} & \cdots & a_{1M} \\ \vdots & \ddots & \cdots \\ a_{M1} & \cdots & a_{MM} \end{bmatrix}$$

$$\frac{d\mathbf{y}(t)}{dt} = \mathbf{A}(t)\mathbf{y}(t) + \mathbf{u}(t)$$

$$\frac{d\mathbf{y}(t)}{dt} = \begin{bmatrix} \frac{dy_1(t)}{dt} \\ \vdots \\ \frac{dy_M(t)}{dt} \end{bmatrix}$$

$$y'_1(t) = a_{11}(t)y_1(t) + a_{12}(t)y_2(t)$$

$$y'_2(t) = a_{21}(t)y_1(t) + a_{22}(t)y_2(t)$$

$$\begin{bmatrix} y'_1(t) \\ y'_2(t) \end{bmatrix} = \begin{bmatrix} a_{11}(t) & a_{12}(t) \\ a_{21}(t) & a_{22}(t) \end{bmatrix} \begin{bmatrix} y_1(t) \\ y_2(t) \end{bmatrix}$$

a homogeneous matrix equation

$$\frac{d\mathbf{y}(t)}{dt} = \mathbf{A}(t)\mathbf{y}(t)$$

a general solution

$$\mathbf{y}(t) = \mathbf{Y}(t)\mathbf{h}$$

$$\mathbf{Y}(t) = \begin{bmatrix} y_{11}(t) & \cdots & y_{1M}(t) \\ \vdots & \ddots & \cdots \\ y_{M1}(t) & \cdots & y_{MM}(t) \end{bmatrix}$$

$$\mathbf{h} = \begin{bmatrix} h_1 \\ \vdots \\ h_M \end{bmatrix}$$

**a fundamental
(solution) matrix**
[nonsingular,
Wronskian positive]

Derivative of matrices

$$\frac{d\mathbf{Y}(t)}{dt} = \begin{bmatrix} \frac{dy_{11}(t)}{dt} & \dots & \frac{dy_{1M}(t)}{dt} \\ \vdots & \ddots & \dots \\ \frac{dy_{M1}(t)}{dt} & \dots & \frac{dy_{MM}(t)}{dt} \end{bmatrix}$$

\mathbf{A} – a constant matrix

$$\frac{d\mathbf{A}}{dt} = \mathbf{0} \quad \boxed{\frac{d(\mathbf{AY}(t))}{dt} = \mathbf{A} \frac{d\mathbf{Y}(t)}{dt}}$$

$$\boxed{\frac{d(\mathbf{Y}_1(t) + \mathbf{Y}_2(t))}{dt} = \frac{d\mathbf{Y}_1(t)}{dt} + \frac{d\mathbf{Y}_2(t)}{dt}}$$

$$\begin{aligned} \frac{d(\mathbf{Y}_1(t)\mathbf{Y}_2(t))}{dt} &= \\ &= \frac{d\mathbf{Y}_1(t)}{dt}\mathbf{Y}_2(t) + \mathbf{Y}_1(t) \frac{d\mathbf{Y}_2(t)}{dt} \end{aligned}$$

$\mathbf{X}_1\mathbf{X}_2 \neq \mathbf{X}_2\mathbf{X}_1$ matrix multiplication
is not commutative

$$\boxed{\frac{d\mathbf{Y}^2(t)}{dt} = \frac{d\mathbf{Y}(t)}{dt}\mathbf{Y}(t) + \mathbf{Y}(t)\frac{d\mathbf{Y}(t)}{dt}}$$

Thus

$$\frac{d\mathbf{Y}^2(t)}{dt} = 2\mathbf{Y}(t) \frac{d\mathbf{Y}(t)}{dt} \text{ only if } \mathbf{Y}(t) \text{ and } \frac{d\mathbf{Y}(t)}{dt} \text{ commute}$$

$$\mathbf{Y}(t)\mathbf{Y}^{-1}(t) = \mathbf{I}$$

$$\frac{d\mathbf{Y}(t)}{dt}\mathbf{Y}^{-1}(t) + \mathbf{Y}(t)\frac{d\mathbf{Y}^{-1}(t)}{dt} = \mathbf{0}$$

$$\boxed{\frac{d\mathbf{Y}^{-1}(t)}{dt} = \mathbf{Y}^{-1}(t) \frac{d\mathbf{Y}(t)}{dt} \mathbf{Y}^{-1}(t)}$$

Normed fundamental matrix

$\leftarrow M$ equations for M solutions

$$\frac{d\mathbf{Y}(t)}{dt} = \mathbf{A}(t)\mathbf{Y}(t) \quad (\&) \quad \frac{d\mathbf{Y}(t)}{dt} = \begin{bmatrix} \frac{dy_{11}(t)}{dt} & \dots & \frac{dy_{1M}(t)}{dt} \\ \vdots & \ddots & \cdots \\ \frac{dy_{M1}(t)}{dt} & \dots & \frac{dy_{MM}(t)}{dt} \end{bmatrix}$$

$$\begin{aligned} \mathbf{Z}(t) &= \Phi(t, t_0)\mathbf{H} \quad \mathbf{H} - \text{nonsingular} \\ \frac{d\mathbf{Z}(t)}{dt} &= \frac{d(\Phi(t, t_0)\mathbf{H})}{dt} = \frac{d(\Phi(t, t_0))}{dt}\mathbf{H} = \\ &= \mathbf{A}(t)\Phi(t, t_0)\mathbf{H} = \mathbf{A}(t)\mathbf{Z}(t) \\ \mathbf{Z}(t) &- \text{a fundamental matrix} \end{aligned}$$

$\mathbf{Y}(t) = \Phi(t, t_0)\mathbf{Y}(t_0)$ – a fundamental matrix with the initial condition $\mathbf{Y}(t_0)$

$$\mathbf{Y}(t) = \Phi(t, t_1)\mathbf{Y}(t_1)$$

$$\mathbf{Y}(t) = \Phi(t, t_1)\Phi(t_1, t_0)\mathbf{Y}(t_0)$$

$$\boxed{\Phi(t, t_0) = \Phi(t, t_1)\Phi(t_1, t_0)}$$

$$\boxed{\Phi^{-1}(t_1, t_0) = \Phi(t_0, t_1)} \quad t = t_0 \text{ above}$$

The solution satisfying the initial cond. $\mathbf{y}(t_0)$
 $\mathbf{y}(t) = \Phi(t, t_0)\mathbf{y}(t_0)$

normed fundamental matrix
(state transition matrix)

$$\Phi(t, t_0) = \mathbf{Y}(t)\mathbf{Y}^{-1}(t_0)$$

$\boxed{\Phi(t_0, t_0) = \mathbf{I}}$ – identity matrix

$\Phi(t, t_0)$ is a solution of $(\&)$ with the initial condition $\mathbf{Y}(t_0) = \mathbf{I}$

Nonhomogeneous equations

$$\frac{dy(t)}{dt} = \mathbf{A}(t)y(t) + \mathbf{u}(t) \quad (\$)$$

Let's try a solution

$$y(t) = y^s(t) + y^0(t)$$

$y^s(t)$ - a particular solution of (\$)

$y^0(t) = \mathbf{Y}(t)\mathbf{h}$ - a general solution
of the homogeneous equation

$$\frac{dy^0(t)}{dt} = \mathbf{A}(t)y^0(t)$$

$$\begin{aligned} \frac{d(y^s(t)+y^0(t))}{dt} &= \mathbf{A}(t)y^s(t) + \\ &+ \mathbf{A}(t)y^0(t) + \mathbf{u}(t) \end{aligned}$$

A general solution of
a nonhomogenous equation =
a particular solution of the
nonhomogenous equation + a general
solution of the homogenous equation

Variation of parameters method

$$y^s(t) = \mathbf{Y}(t)\mathbf{h}(t) \quad \mathbf{Y}(t) - \text{a fundamental matrix}$$

$$\cancel{\frac{d\mathbf{Y}(t)}{dt}\mathbf{h}(t) + \mathbf{Y}(t)\frac{d\mathbf{h}(t)}{dt}} = \cancel{\mathbf{A}(t)\mathbf{Y}(t)\mathbf{h}(t) + \mathbf{u}(t)}$$

$$\mathbf{Y}(t)\frac{d\mathbf{h}(t)}{dt} = \mathbf{u}(t)$$

$$\frac{d\mathbf{h}(t)}{dt} = \mathbf{Y}^{-1}(t)\mathbf{u}(t)$$

$$\mathbf{h}(t) = \int_{t_0}^t \mathbf{Y}^{-1}(\tau)\mathbf{u}(\tau)d\tau \quad \mathbf{h}(t_0) = \mathbf{0}$$

$$y^s(t) = \mathbf{Y}(t) \int_{t_0}^t \mathbf{Y}^{-1}(\tau)\mathbf{u}(\tau)d\tau$$

$$y(t) = \mathbf{Y}(t)\mathbf{h} + \mathbf{Y}(t) \int_{t_0}^t \mathbf{Y}^{-1}(\tau)\mathbf{u}(\tau)d\tau$$

General solution of the nonhomogenous equation

General solution of nonhomogeneous equations

$$\mathbf{y}^s(t) = \mathbf{Y}(t) \int_{t_0}^t \mathbf{Y}^{-1}(\tau) \mathbf{u}(\tau) d\tau$$

$$\mathbf{y}(t) = \mathbf{Y}(t) \mathbf{h} + \mathbf{Y}(t) \int_{t_0}^t \mathbf{Y}^{-1}(\tau) \mathbf{u}(\tau) d\tau$$

$$\begin{aligned}\mathbf{y}(t) &= \mathbf{Y}(t) \mathbf{Y}^{-1}(t_0) \mathbf{Y}(t_0) \mathbf{h} + \\ &+ \mathbf{Y}(t) \mathbf{Y}^{-1}(t_0) \int_{t_0}^t (\mathbf{Y}(\tau) \mathbf{Y}^{-1}(t_0))^{-1} \mathbf{u}(\tau) d\tau =\end{aligned}$$

$$= \Phi(t, t_0) \mathbf{b} + \Phi(t, t_0) \int_{t_0}^t \Phi^{-1}(\tau, t_0) \mathbf{u}(\tau) d\tau$$

$$\boxed{\mathbf{y}(t) = \Phi(t, t_0) \mathbf{b} + \int_{t_0}^t \Phi(t, \tau) \mathbf{u}(\tau) d\tau}$$

$\Phi(t, \tau)$ or $\mathbf{Y}(t)$ has to be known!

$$\mathbf{b} = \mathbf{Y}(t_0) \mathbf{h}$$

$$\frac{dy(t)}{dt} = a(t)y(t) \quad y(t_0) = y_0$$

$$\frac{dy}{y} = a(t) dt \quad y(t) = y_0 e^{\int_{t_0}^t a(\tau) d\tau}$$

A question

$$\frac{d\mathbf{y}(t)}{dt} = \mathbf{A}(t)\mathbf{y}(t)$$

Is it true that $\mathbf{y}(t) = e^{\int_{t_0}^t \mathbf{A}(\tau) d\tau} \mathbf{y}_0$?

$$\int \mathbf{A} dt = \begin{bmatrix} \int a_{11}(t) dt & \dots & \int a_{1M}(t) dt \\ \vdots & \ddots & \vdots \\ \int a_{M1}(t) dt & \dots & \int a_{MM}(t) dt \end{bmatrix}$$

true only if the matrix and its integral commute

$$\mathbf{A}(t) \int \mathbf{A}(\tau) d\tau = \int \mathbf{A}(\tau) d\tau \mathbf{A}(t)$$

Computer Modeling

Ordinary differential equations
continued

$$e^a = \lim_{n \rightarrow \infty} \left(1 + \frac{a}{n}\right)^n$$

Function $e^{\mathbf{A}t}$

$$e^a = \sum_{i=0}^{\infty} \frac{a^i}{i!}$$

$$e^{\mathbf{A}t} = \mathbf{I} + \mathbf{A}t + \frac{(\mathbf{A}t)^2}{2!} + \dots = \sum_{i=0}^{\infty} \frac{\mathbf{A}^i t^i}{i!}$$

M^2 power series – convergent

$\mathbf{A}(t)$ continuous on $b \leq t \leq d$, then the series converges uniformly

$$\mathbf{A} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} = \mathbf{0} \quad e^{\mathbf{A}t} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\mathbf{A} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \mathbf{I} \quad e^{\mathbf{A}t} = e^t \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\frac{de^{\mathbf{A}t}}{dt} = \mathbf{A} + \mathbf{A}^2 t + \mathbf{A}^3 \frac{t^2}{2!} + \dots = \mathbf{A} e^{\mathbf{A}t}$$

1 \mathbf{S} nonsingular

$$e^{\mathbf{SAS}^{-1}} = \mathbf{S} e^{\mathbf{A}} \mathbf{S}^{-1} \quad (\text{from the definition})$$

$$\boxed{\mathbf{I} = \mathbf{SIS}^{-1}}$$

$$\boxed{(\mathbf{SAS}^{-1})\mathbf{SAS}^{-1} = \mathbf{SA}^2\mathbf{S}^{-1}}$$

(proof from the definition)

$$\boxed{e^{\mathbf{A}_1} e^{\mathbf{A}_2} = e^{\mathbf{A}_2} e^{\mathbf{A}_1} \quad \text{if } \mathbf{A}_1, \mathbf{A}_2 \text{ commute}}$$

2 $\mathbf{A}_1, \mathbf{A}_2$ commute ($\mathbf{A}_1 \mathbf{A}_2 = \mathbf{A}_2 \mathbf{A}_1$)

$$e^{\mathbf{A}_1 + \mathbf{A}_2} = e^{\mathbf{A}_1} e^{\mathbf{A}_2}$$

corrolaries

$$e^{\mathbf{A}(t_1 + t_2)} = e^{\mathbf{At}_1} e^{\mathbf{At}_2} \quad (\mathbf{At}_1, \mathbf{At}_2 \text{ commute})$$

$$(e^{\mathbf{A}})^{-1} = e^{-\mathbf{A}} \quad (t_1 = 1, t_2 = -1)$$

$$3 \mathbf{A} = \begin{bmatrix} \mathbf{A}_1 & & 0 \\ & \ddots & \\ 0 & & \mathbf{A}_m \end{bmatrix} \quad e^{\mathbf{A}} = \begin{bmatrix} e^{\mathbf{A}_1} & & 0 \\ & \ddots & \\ 0 & & e^{\mathbf{A}_m} \end{bmatrix}$$

because

$$\mathbf{A}^i = \begin{bmatrix} \mathbf{A}_1^i & & 0 \\ & \ddots & \\ 0 & & \mathbf{A}_m^i \end{bmatrix}$$

4 $\mathbf{A}(t), \frac{d\mathbf{A}(t)}{dt}$ commute ($\mathbf{A}(t) \frac{d\mathbf{A}(t)}{dt} = \frac{d\mathbf{A}(t)}{dt} \mathbf{A}(t)$)

$$\frac{de^{\mathbf{A}(t)}}{dt} = \frac{d\mathbf{A}(t)}{dt} e^{\mathbf{A}(t)}$$

$$\boxed{\det e^{\mathbf{A}} = e^{\text{tr A}} > 0}$$

$$\text{tr A} = a_{11} + \dots + a_{MM}$$

A fundamental matrix

$$e^{\int_{t_0}^t \mathbf{A}(\tau) d\tau} ?$$

$$\frac{dY(t)}{dt} = \mathbf{A}(t)Y(t) \quad Y(0) = \mathbf{I}$$

Is it true that $Y(t) = e^{\int_{t_0}^t \mathbf{A}(\tau) d\tau}$?

$$e^{\int_{t_0}^t \mathbf{A}(\tau) d\tau} = \mathbf{I} + \int_{t_0}^t \mathbf{A}(\tau) d\tau + \\ + \frac{1}{2!} \left(\int_{t_0}^t \mathbf{A}(\tau) d\tau \right)^2 + \frac{1}{3!} \left(\int_{t_0}^t \mathbf{A}(\tau) d\tau \right)^3 \dots$$

$$\frac{d}{dt} e^{\int_{t_0}^t \mathbf{A}(\tau) d\tau} = \mathbf{A}(t) + \frac{1}{2!} \mathbf{A}(t) \int_{t_0}^t \mathbf{A}(\tau) d\tau + \\ + \frac{1}{2!} \int_{t_0}^t \mathbf{A}(\tau) d\tau \mathbf{A}(t) + \frac{1}{3!} \mathbf{A}(t) \left(\int_{t_0}^t \mathbf{A}(\tau) d\tau \right)^2 + \\ + \frac{1}{3!} \int_{t_0}^t \mathbf{A}(\tau) d\tau \mathbf{A}(t) \int_{t_0}^t \mathbf{A}(\tau) d\tau + \\ + \frac{1}{3!} \left(\int_{t_0}^t \mathbf{A}(\tau) d\tau \right)^2 \mathbf{A}(t) + \dots =$$

If $\mathbf{A}(t) \int_{t_0}^t \mathbf{A}(\tau) d\tau = \int_{t_0}^t \mathbf{A}(\tau) d\tau \mathbf{A}(t)$
(commute)

$$\frac{d}{dt} e^{\int_{t_0}^t \mathbf{A}(\tau) d\tau} = \mathbf{A}(t) \left[\mathbf{I} + \int_{t_0}^t \mathbf{A}(\tau) d\tau + \frac{1}{2!} \left(\int_{t_0}^t \mathbf{A}(\tau) d\tau \right)^2 + \frac{1}{3!} \left(\int_{t_0}^t \mathbf{A}(\tau) d\tau \right)^3 \dots \right] = \\ = \mathbf{A}(t) e^{\int_{t_0}^t \mathbf{A}(\tau) d\tau}$$

In this case, $Y(t) = e^{\int_{t_0}^t \mathbf{A}(\tau) d\tau}$ (nonsingular)
is a fundamental matrix, and it is
the fundamental matrix normed at t_0 , i.e.

$$\Phi(t, t_0) = e^{\int_{t_0}^t \mathbf{A}(\tau) d\tau}$$

Example $\mathbf{A}(t) = a(t) \mathbf{B}$ \mathbf{B} – a constant matrix

Equations with constant coefficients

commute

$$\frac{d\mathbf{y}(t)}{dt} = \mathbf{A}\mathbf{y}(t)$$

$$\mathbf{y}(t) = e^{\mathbf{A}(t-t_0)}\mathbf{y}_0$$

$$\mathbf{A} \int \mathbf{A} dt = \int \mathbf{A} dt \mathbf{A}$$

$e^{\mathbf{A}(t-t_0)}$ the normed fundamental matrix
for $t = t_0$ ($= \mathbf{I}$)

any fundamental matrix $\mathbf{Y}(t)$ satisfies

$$\frac{d\mathbf{Y}(t)}{dt} = \mathbf{A}\mathbf{Y}(t)$$

$$\frac{d\mathbf{y}(t)}{dt} = \mathbf{A}\mathbf{y}(t) + \mathbf{u}(t)$$

a general solution

$$\mathbf{y}(t) = e^{\mathbf{A}(t-t_0)}\mathbf{h} + \int_{t_0}^t e^{\mathbf{A}(t-\tau)}\mathbf{u}(\tau)d\tau$$

a particular solution for $\mathbf{y}(t_0) = \mathbf{y}_0$

$$\mathbf{y}(t) = e^{\mathbf{A}(t-t_0)}\mathbf{y}_0 + \int_{t_0}^t e^{\mathbf{A}(t-\tau)}\mathbf{u}(\tau)d\tau$$

$$\frac{d\mathbf{Y}(t)}{dt} = \mathbf{A}\mathbf{Y}(t) \quad \mathbf{Y}(0) = \mathbf{I}$$

$$s\mathcal{L}\{\mathbf{Y}(t)\} - \mathbf{I} = \mathbf{A}\mathcal{L}\{\mathbf{Y}(t)\} \quad \mathcal{L}\{\mathbf{Y}(t)\} = \mathbf{Y}(s)$$

$$(s\mathbf{I} - \mathbf{A})\mathcal{L}\{\mathbf{Y}(t)\} = \mathbf{I}$$

$$\mathcal{L}\{\mathbf{Y}(t)\} = (s\mathbf{I} - \mathbf{A})^{-1}$$

$(s\mathbf{I} - \mathbf{A})$ nonsingular
generically]

$$e^{\mathbf{At}} = \mathcal{L}^{-1}\{(s\mathbf{I} - \mathbf{A})^{-1}\}$$

$$s\mathbf{y}(s) = \mathbf{A}\mathbf{y}(s) + \mathbf{u}(s) \quad (\mathbf{y}(t_0) = \mathbf{0})$$

$$(s\mathbf{I} - \mathbf{A})\mathbf{y}(s) = \mathbf{u}(s)$$

$$\mathbf{y}(s) = (s\mathbf{I} - \mathbf{A})^{-1}\mathbf{u}(s)$$

$\mathbf{G}(s) = (s\mathbf{I} - \mathbf{A})^{-1}$ the transfer matrix

$$\mathbf{G}(s) = \mathcal{L}\{\mathbf{Y}(t)\}$$

Nonsingularity of $e^{\mathbf{A}}$

$$\det e^{\mathbf{A}} = e^{\text{tr}\mathbf{A}}$$

$$\text{tr } \mathbf{A} = a_{11} + \dots + a_{MM}$$

Sketch of a proof

$$\frac{d\mathbf{Z}(t)}{dt} = \mathbf{A}\mathbf{Z}(t), \quad \mathbf{Z}(0) = \mathbf{I}$$

Solution: $\mathbf{Z}(t) = e^{\mathbf{At}}$ a fundamental matrix

$$M=2$$

$$\begin{bmatrix} \frac{dz_{11}}{dt} & \frac{dz_{12}}{dt} \\ \frac{dz_{21}}{dt} & \frac{dz_{22}}{dt} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} z_{11} & z_{12} \\ z_{21} & z_{22} \end{bmatrix} = \\ = \begin{bmatrix} a_{11}z_{11} + a_{12}z_{21} & a_{11}z_{12} + a_{12}z_{22} \\ a_{21}z_{11} + a_{22}z_{21} & a_{21}z_{12} + a_{22}z_{22} \end{bmatrix}$$

$$\det e^{\mathbf{At}} = \det \mathbf{Z}(t)$$

$$\begin{aligned} \frac{d}{dt} \begin{vmatrix} z_{11} & z_{12} \\ z_{21} & z_{22} \end{vmatrix} &= \frac{d}{dt} (z_{11}z_{22} - z_{12}z_{21}) = \\ &= \frac{dz_{11}}{dt} z_{22} + z_{11} \frac{dz_{22}}{dt} - \frac{dz_{12}}{dt} z_{21} - z_{12} \frac{dz_{21}}{dt} = \\ &= \begin{vmatrix} \frac{dz_{11}}{dt} & \frac{dz_{12}}{dt} \\ z_{21} & z_{22} \end{vmatrix} + \begin{vmatrix} z_{11} & z_{12} \\ \frac{dz_{21}}{dt} & \frac{dz_{22}}{dt} \end{vmatrix} = \\ &= \begin{vmatrix} a_{11}z_{11} + a_{12}z_{21} & a_{11}z_{12} + a_{12}z_{22} \\ z_{21} & z_{22} \end{vmatrix} + \\ &\quad + \begin{vmatrix} z_{11} & z_{12} \\ a_{21}z_{11} + a_{22}z_{21} & a_{21}z_{12} + a_{22}z_{22} \end{vmatrix} = \\ &= \begin{vmatrix} a_{11}z_{11} & a_{11}z_{12} \\ z_{21} & z_{22} \end{vmatrix} + \begin{vmatrix} z_{11} & z_{12} \\ a_{22}z_{21} & a_{22}z_{22} \end{vmatrix} = \\ &= (a_{11} + a_{22}) \det \mathbf{Z} = \text{tr } \mathbf{A} \det \mathbf{Z} \\ \frac{d \det \mathbf{Z}(t)}{dt} &= \text{tr } \mathbf{A} \det \mathbf{Z}(t) \quad \det \mathbf{Z}(0) = 1 \\ \det \mathbf{Z}(t) &= e^{\text{tr}\mathbf{A} t} \quad t = 1 \\ \det e^{\mathbf{A}} &= e^{\text{tr}\mathbf{A}} \end{aligned}$$

Nonsingularity of $e^{\mathbf{A}}$ cont.

$$M = 3$$

$$\begin{bmatrix} \frac{dz_{11}}{dt} & \frac{dz_{12}}{dt} & \frac{dz_{13}}{dt} \\ \frac{dz_{21}}{dt} & \frac{dz_{22}}{dt} & \frac{dz_{23}}{dt} \\ \frac{dz_{31}}{dt} & \frac{dz_{32}}{dt} & \frac{dz_{33}}{dt} \end{bmatrix} =$$

$$= \begin{bmatrix} \sum_{i=1}^3 a_{1i} z_{i1} & \sum_{i=1}^3 a_{1i} z_{i2} & \sum_{i=1}^3 a_{1i} z_{i3} \\ \sum_{i=1}^3 a_{2i} z_{i1} & \sum_{i=1}^3 a_{2i} z_{i2} & \sum_{i=1}^3 a_{2i} z_{i3} \\ \sum_{i=1}^3 a_{3i} z_{i1} & \sum_{i=1}^3 a_{3i} z_{i2} & \sum_{i=1}^3 a_{3i} z_{i3} \end{bmatrix}$$

Laplace expansion along the 1st row

$$\begin{vmatrix} z_{11} & z_{12} & z_{13} \\ z_{21} & z_{22} & z_{23} \\ z_{31} & z_{32} & z_{33} \end{vmatrix} = z_{11} \begin{vmatrix} z_{22} & z_{23} \\ z_{32} & z_{33} \end{vmatrix} -$$

$$- z_{12} \begin{vmatrix} z_{21} & z_{13} \\ z_{31} & z_{33} \end{vmatrix} + z_{13} \begin{vmatrix} z_{21} & z_{22} \\ z_{31} & z_{32} \end{vmatrix}$$

$$\begin{aligned} \frac{d}{dt} \begin{vmatrix} z_{11} & z_{12} & z_{13} \\ z_{21} & z_{22} & z_{23} \\ z_{31} & z_{32} & z_{33} \end{vmatrix} &= \frac{dz_{11}}{dt} \begin{vmatrix} z_{22} & z_{23} \\ z_{32} & z_{33} \end{vmatrix} + \\ z_{11} \frac{d}{dt} \begin{vmatrix} z_{22} & z_{23} \\ z_{32} & z_{33} \end{vmatrix} &- \frac{dz_{12}}{dt} \begin{vmatrix} z_{21} & z_{23} \\ z_{31} & z_{33} \end{vmatrix} - \\ - z_{12} \frac{d}{dt} \begin{vmatrix} z_{21} & z_{13} \\ z_{31} & z_{33} \end{vmatrix} &+ \frac{dz_{13}}{dt} \begin{vmatrix} z_{21} & z_{22} \\ z_{31} & z_{32} \end{vmatrix} + \\ + z_{13} \frac{d}{dt} \begin{vmatrix} z_{21} & z_{22} \\ z_{31} & z_{32} \end{vmatrix} &= \begin{vmatrix} \frac{dz_{11}}{dt} & \frac{dz_{12}}{dt} & \frac{dz_{13}}{dt} \\ z_{21} & z_{22} & z_{23} \\ z_{31} & z_{32} & z_{33} \end{vmatrix} + \end{aligned}$$

$$\begin{aligned} + \begin{vmatrix} z_{11} & z_{12} & z_{13} \\ \frac{dz_{21}}{dt} & \frac{dz_{22}}{dt} & \frac{dz_{23}}{dt} \\ z_{31} & z_{32} & z_{33} \end{vmatrix} + \begin{vmatrix} z_{11} & z_{12} & z_{13} \\ z_{21} & z_{22} & z_{23} \\ \frac{dz_{31}}{dt} & \frac{dz_{32}}{dt} & \frac{dz_{33}}{dt} \end{vmatrix} &= \\ = \begin{vmatrix} a_{11}z_{11} & a_{11}z_{12} & a_{11}z_{13} \\ z_{21} & z_{22} & z_{23} \\ z_{31} & z_{32} & z_{33} \end{vmatrix} + \dots &= \\ = (a_{11} + a_{22} + a_{33}) \det \mathbf{Z} &= \text{tr } \mathbf{A} \det \mathbf{Z} \end{aligned}$$

and further as before

Jordan canonical form: distinct eigenvalues

$\frac{d\mathbf{Y}(t)}{dt} = \mathbf{A}\mathbf{Y}(t)$, transformation of $\mathbf{Y}(t)$ to

$\mathbf{Z}(t) = \mathbf{S}^{-1}\mathbf{Y}(t)$ \mathbf{S} - a nonsingular matrix

$\frac{d\mathbf{Z}(t)}{dt} = \mathbf{\Lambda}\mathbf{Z}(t)$ $\mathbf{\Lambda} = \mathbf{S}^{-1}\mathbf{A}\mathbf{S}$

$\mathbf{\Lambda}$ - Jordan canonical form $\mathbf{Z}(t) = e^{\mathbf{\Lambda}t}$

The form depends on eigenvalues of \mathbf{A}

$(\mathbf{A} - \lambda\mathbf{I})\mathbf{x} = \mathbf{0}$ the characteristic equation

$\det(\mathbf{A} - \lambda\mathbf{I}) = 0 \rightarrow M$ eigenvalues λ_i

\mathbf{x}_i - eigenvectors

distinct (single) eigenvalues:

columns of \mathbf{S} are eigenvectors, as

$\mathbf{AS} = \mathbf{S}\mathbf{\Lambda}$

$\mathbf{A}[\mathbf{s}_1, \dots, \mathbf{s}_M] = [\lambda_1 \mathbf{s}_1, \dots, \lambda_M \mathbf{s}_M]$

$(\mathbf{A} - \lambda_1 \mathbf{I})\mathbf{s}_1 = \mathbf{0}, \dots, (\mathbf{A} - \lambda_M \mathbf{I})\mathbf{s}_M = \mathbf{0}$

For distinct eigenvalues $\lambda_i \neq \lambda_j$ for $i \neq j$

$$\mathbf{\Lambda} = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_M \end{bmatrix} \quad e^{\mathbf{\Lambda}t} = \begin{bmatrix} e^{\lambda_1 t} & & 0 \\ & \ddots & \\ 0 & & e^{\lambda_M t} \end{bmatrix}$$

$$\mathbf{Y}(t) = \mathbf{SZ}(t) = \begin{bmatrix} s_{11}e^{\lambda_1 t} & \dots & s_{1M}e^{\lambda_M t} \\ \vdots & \ddots & \vdots \\ s_{M1}e^{\lambda_1 t} & \dots & s_{MM}e^{\lambda_M t} \end{bmatrix}$$

$\mathbf{Y}(t)$ - a fundamental matrix (**Example (E)**)

$\mathbf{s}_i = \begin{bmatrix} s_{i1} \\ \vdots \\ s_{iM} \end{bmatrix} \neq \mathbf{0}$

a general solution
of a nonhomogeneous equation
for distinct eigenvalues

$$\mathbf{y}^0(t) = \mathbf{Y}(t)\mathbf{h} = \sum_{i=1}^M h_i \mathbf{s}_i e^{\lambda_i t} = \sum_{i=1}^M \mathbf{c}_i e^{\lambda_i t}$$

An example (1)

$$\frac{dy_1(t)}{dt} = 5y_1(t) + 4y_2(t)$$

$$\frac{dy_2(t)}{dt} = 4y_1(t) + 5y_2(t)$$

the characteristic equation

$$\begin{vmatrix} 5 - \lambda & 4 \\ 4 & 5 - \lambda \end{vmatrix} = \lambda^2 - 10\lambda + 9 = 0$$

$$\lambda_1 = 1, \quad \lambda_2 = 9$$

an eigenvector for $\lambda_1 = 1$

$$\begin{bmatrix} 4 & 4 \\ 4 & 4 \end{bmatrix} \begin{bmatrix} s_{11} \\ s_{21} \end{bmatrix} = \mathbf{0}, \quad s_{11} = 1, s_{21} = -1$$

an eigenvector for $\lambda_2 = 9$

$$\begin{bmatrix} -4 & 4 \\ 4 & -4 \end{bmatrix} \begin{bmatrix} s_{12} \\ s_{22} \end{bmatrix} = \mathbf{0}, \quad s_{12} = 1, s_{22} = 1$$

$$W(t) = \begin{vmatrix} e^t & e^{9t} \\ -e^t & e^{9t} \end{vmatrix} = 2e^{10t} > 0$$

a fundamental matrix

$$Y(t) = \begin{bmatrix} e^t & e^{9t} \\ -e^t & e^{9t} \end{bmatrix}$$

a general solution

$$y_1(t) = h_1 e^t + h_2 e^{9t}$$

$$y_2(t) = -h_1 e^t + h_2 e^{9t}$$

checking

$$L_1: h_1 e^t + 9h_2 e^{9t}$$

$$P_1: 5(h_1 e^t + h_2 e^{9t}) + 4(-h_1 e^t + h_2 e^{9t}) = L_1$$

$$L_2: -h_1 e^t + 9h_2 e^{9t}$$

$$P_2: 4(h_1 e^t + h_2 e^{9t}) + 5(-h_1 e^t + h_2 e^{9t}) = L_2$$

An example (2)

$$\frac{dy_1(t)}{dt} = y_1(t) - y_2(t)$$

$$\frac{dy_2(t)}{dt} = y_1(t) + y_2(t)$$

the characteristic equation

$$\begin{vmatrix} 1-\lambda & -1 \\ 1 & 1-\lambda \end{vmatrix} = \lambda^2 - 2\lambda + 2 = 0$$

$$\lambda_1 = 1+i, \quad \lambda_2 = 1-i$$

an eigenvector for $\lambda_1 = 1+i$

$$\begin{bmatrix} -i & -1 \\ 1 & -i \end{bmatrix} \begin{bmatrix} s_{11} \\ s_{21} \end{bmatrix} = \mathbf{0}, \quad s_{11} = i, s_{21} = 1$$

an eigenvector for $\lambda_2 = 1-i$

$$\begin{bmatrix} i & -1 \\ 1 & i \end{bmatrix} \begin{bmatrix} s_{12} \\ s_{22} \end{bmatrix} = \mathbf{0}, \quad s_{12} = i, s_{22} = -1$$

$$y_{11}(t) = ie^{(i+1)t} = -e^t \sin t + ie^t \cos t$$

$$y_{21}(t) = e^{(i+1)t} = e^t \cos t + ie^t \sin t$$

$$y_{12}(t) = ie^{(i-1)t} = e^t \sin t + ie^t \cos t$$

$$y_{22}(t) = -e^{(i-1)t} = -e^t \cos t + ie^t \sin t$$

$$W(t) = \begin{vmatrix} e^t \cos t & -e^t \sin t \\ e^t \sin t & e^t \cos t \end{vmatrix} = e^{2t} > 0$$

a fundamental matrix

$$Y(t) = \begin{bmatrix} e^t \cos t & -e^t \sin t \\ e^t \sin t & e^t \cos t \end{bmatrix}$$

a general solution

$$y_1(t) = h_1 e^t \cos t - h_2 e^t \sin t$$

$$y_2(t) = h_1 e^t \sin t + h_2 e^t \cos t$$

$$\Lambda = \mathbf{S}^{-1} \mathbf{A} \mathbf{S}$$

Jordan canonical form: multiple eigenvalues

$$\Lambda = \begin{bmatrix} \Lambda_1 & & \mathbf{0} \\ & \ddots & \\ \mathbf{0} & & \Lambda_L \end{bmatrix}$$

L – number
of different
eigenvalues

$$\Lambda_l = \begin{bmatrix} \Lambda_{l1} & & \mathbf{0} \\ & \ddots & \\ \mathbf{0} & & \Lambda_{lk_l} \end{bmatrix}$$

k_l - multiplicity of
the l -th eigenvalue

$$\Lambda_{lj} = \begin{bmatrix} \lambda_l & 0 & 0 & \cdots & 0 & 0 \\ 1 & \lambda_l & 0 & \cdots & 0 & 0 \\ 0 & 1 & \lambda_l & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 & \lambda_l \end{bmatrix}$$

a Jordan
block

$$\Lambda_{lj} = \lambda_l \mathbf{I} + \mathbf{F}_{lj}$$

$$\mathbf{F}_{lj} = \begin{bmatrix} 0 & 0 & \cdots & 0 \\ 1 & 0 & \cdots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

$$e^{\Lambda t} = \begin{bmatrix} e^{\Lambda_{11}t} & & \mathbf{0} \\ & \ddots & \\ \mathbf{0} & & e^{\Lambda_{Lk_L}t} \end{bmatrix}$$

$$e^{\Lambda_{lj}t} = e^{(\lambda_l \mathbf{I} + \mathbf{F}_{lj})t} = e^{\lambda_l t} e^{\mathbf{F}_{lj}t}$$

$$e^{\mathbf{F}_{lj}t} = \mathbf{I} + \mathbf{F}_{lj}t + \frac{1}{2!} \mathbf{F}_{lj}^2 t^2 + \dots$$

assume $k_l = 3$

$$\mathbf{F}_{lj} = \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \quad \mathbf{F}_{lj}^2 = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \end{bmatrix} \quad \mathbf{F}_{lj}^3 = \mathbf{0}$$

$$e^{\Lambda_{lj}t} = e^{\lambda_l t} (\mathbf{I} + \mathbf{F}_{lj}t + \frac{1}{2!} \mathbf{F}_{lj}^2 t^2) = e^{\lambda_l t} \begin{bmatrix} 1 & 0 & 0 \\ t & 1 & 0 \\ \frac{1}{2!} t^2 & t & 1 \end{bmatrix}$$

in general

$$\mathbf{y}(t) = \sum_{l=1}^L \sum_{j=0}^{k_l-1} t^j e^{\lambda_l t} \mathbf{z}_{lj} \mathbf{h}$$

\mathbf{z}_{lj} depends only on \mathbf{A} , not on t

\mathbf{I} and \mathbf{F}_{lj}
commute

Jordan canonical form: multiple eigenvalues

$$\begin{aligned}
 \mathbf{y}(t) &= \sum_{l=1}^L \sum_{j=0}^{k_l-1} t^j e^{\lambda_l t} \mathbf{Z}_{lj} \mathbf{h} = \\
 &= e^{\lambda_1 t} \mathbf{c}_{10} + t e^{\lambda_1 t} \mathbf{c}_{11} + \cdots + t^{k_1-1} e^{\lambda_1 t} \mathbf{c}_{1,k_1-1} + \\
 &\quad \dots \\
 &+ e^{\lambda_L t} \mathbf{c}_{L0} + t e^{\lambda_L t} \mathbf{c}_{L1} + \cdots + t^{k_L-1} e^{\lambda_L t} \mathbf{c}_{L,k_L-1}
 \end{aligned}$$

Jordan decomposition is numerically unstable

Example

$$\mathbf{A} = \begin{bmatrix} 1 & \varepsilon \\ 1 & 1 \end{bmatrix}$$

For $\varepsilon = 0$ the Jordan form

$$\mathbf{A} = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}$$

For $\varepsilon > 0$

$$\begin{vmatrix} 1-\lambda & \varepsilon \\ -1 & -\lambda \end{vmatrix} = \lambda^2 - 2\lambda + 1 - \varepsilon = 0$$

$$\lambda_1 = 1 - \sqrt{\varepsilon}, \quad \lambda_2 = 1 + \sqrt{\varepsilon}$$

$$\begin{bmatrix} \sqrt{\varepsilon} & \varepsilon \\ 1 & \sqrt{\varepsilon} \end{bmatrix} \begin{bmatrix} s_{11} \\ s_{21} \end{bmatrix} = 0 \quad \begin{aligned} s_{11} &= -\sqrt{\varepsilon} \\ s_{21} &= 1 \end{aligned}$$

$$\begin{bmatrix} -\sqrt{\varepsilon} & \varepsilon \\ 1 & -\sqrt{\varepsilon} \end{bmatrix} \begin{bmatrix} s_{12} \\ s_{22} \end{bmatrix} = 0 \quad \begin{array}{l} s_{12} = 1 \\ s_{22} = \sqrt{\varepsilon} \end{array}$$

$$\mathbf{S} = \begin{bmatrix} -\sqrt{\varepsilon} & 1 \\ 1 & \sqrt{\varepsilon} \end{bmatrix} \quad \mathbf{S}^{-1} = \begin{bmatrix} -\frac{1}{2\sqrt{\varepsilon}} & \frac{1}{2} \\ \frac{1}{2\sqrt{\varepsilon}} & \frac{1}{2} \end{bmatrix}$$

$$\Lambda = \mathbf{S}^{-1} \mathbf{A} \mathbf{S} = \begin{bmatrix} -\frac{1}{2\sqrt{\varepsilon}} & \frac{1}{2} \\ \frac{1}{2\sqrt{\varepsilon}} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 1 & \varepsilon \\ 1 & 1 \end{bmatrix} \begin{bmatrix} -\sqrt{\varepsilon} & 1 \\ 1 & \sqrt{\varepsilon} \end{bmatrix} =$$

$$= \begin{bmatrix} 1 - \sqrt{\varepsilon} & 0 \\ 0 & 1 + \sqrt{\varepsilon} \end{bmatrix}$$

Calculation of $e^{\mathbf{A}t}$

from the definition (1)

$$e^{\mathbf{A}t} = \mathbf{I} + \mathbf{A}t + \frac{\mathbf{A}^2 t^2}{2!} + \cdots = \sum_{i=0}^{\infty} \frac{\mathbf{A}^i t^i}{i!}$$

1. $\mathbf{A} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$ $e^{\mathbf{A}t} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

Hints:

2. $\mathbf{A} = \begin{bmatrix} 0 & b \\ 0 & 0 \end{bmatrix}$ $e^{\mathbf{A}t} = \begin{bmatrix} 1 & bt \\ 0 & 1 \end{bmatrix}$

2. $\mathbf{A}^2 = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$

3. $\mathbf{A} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ $e^{\mathbf{A}t} = e^t \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

3. $\mathbf{A}^i = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

4. $\mathbf{A} = \begin{bmatrix} a & 0 \\ 0 & a \end{bmatrix}$ $e^{\mathbf{A}t} = e^{at} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

4. $\mathbf{A}^i = \begin{bmatrix} a^i & 0 \\ 0 & a^i \end{bmatrix}$

5. $\mathbf{A} = \begin{bmatrix} a & b \\ 0 & a \end{bmatrix}$ $e^{\mathbf{A}t} = e^{at} \begin{bmatrix} 1 & bt \\ 0 & 1 \end{bmatrix}$

5. $\mathbf{A} = \begin{bmatrix} a & 0 \\ 0 & a \end{bmatrix} + \begin{bmatrix} 0 & b \\ 0 & 0 \end{bmatrix}$ (commute)

$$e^{\mathbf{A}_1 + \mathbf{A}_2} = e^{\mathbf{A}_1} e^{\mathbf{A}_2} \quad (\mathbf{A}_1 \mathbf{A}_2 = \mathbf{A}_2 \mathbf{A}_1)$$

$$\begin{bmatrix} a & 0 \\ 0 & a \end{bmatrix} \begin{bmatrix} 0 & b \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & b \\ 0 & 0 \end{bmatrix} \begin{bmatrix} a & 0 \\ 0 & a \end{bmatrix} = \begin{bmatrix} 0 & ab \\ 0 & 0 \end{bmatrix}$$

Calculation of $e^{\mathbf{At}}$

from the definition (2)

$$e^{\mathbf{At}} = \mathbf{I} + \mathbf{At} + \frac{\mathbf{A}^2 t^2}{2!} + \dots = \sum_{i=0}^{\infty} \frac{\mathbf{A}^i t^i}{i!}$$

6. $\mathbf{A} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad e^{\mathbf{At}} = \begin{bmatrix} \cosh t & \sinh t \\ \sinh t & \cosh t \end{bmatrix}$

$$\mathbf{A}^2 = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \mathbf{I}$$

$$\mathbf{A}^i = \begin{cases} \mathbf{A} & \text{for } i \text{ odd} \\ \mathbf{I} & \text{for } i \text{ even} \end{cases}$$

$$\sum_{i=0}^{\infty} \frac{\mathbf{A}^i t^i}{i!} = \mathbf{I} + \mathbf{At} + \frac{\mathbf{I}t^2}{2!} + \frac{\mathbf{At}^3}{2!} + \dots =$$

$$= \begin{bmatrix} 1 + \frac{t^2}{2!} + \frac{t^4}{4!} + \dots & t + \frac{t^3}{3!} + \frac{t^5}{5!} + \dots \\ t + \frac{t^3}{3!} + \frac{t^5}{5!} + \dots & 1 + \frac{t^2}{2!} + \frac{t^4}{4!} + \dots \end{bmatrix} =$$

$$= \begin{bmatrix} \frac{1}{2}(e^t + e^{-t}) & \frac{1}{2}(e^t - e^{-t}) \\ \frac{1}{2}(e^t - e^{-t}) & \frac{1}{2}(e^t + e^{-t}) \end{bmatrix} =$$

$$= \begin{bmatrix} \cosh t & \sinh t \\ \sinh t & \cosh t \end{bmatrix}$$

$$e^t + e^{-t} = 1 + t + \frac{t^2}{2!} + \dots \quad 1 - t + \frac{t^2}{2!} + \dots$$

7. $\mathbf{B} = \begin{bmatrix} 0 & b \\ b & 0 \end{bmatrix} \quad e^{\mathbf{B}t} = \begin{bmatrix} \cosh bt & \sinh bt \\ \sinh bt & \cosh bt \end{bmatrix}$

$$\mathbf{B} = \begin{bmatrix} 0 & b \\ b & 0 \end{bmatrix} = b \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = b\mathbf{A}$$

$$\mathbf{B}^2 = b^2 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = b^2 \mathbf{I} \quad \mathbf{B}^3 = b^3 \mathbf{A}$$

$$\mathbf{B}^i = \begin{cases} b^i \mathbf{A} & \text{for } i \text{ odd} \\ b^i \mathbf{I} & \text{for } i \text{ even} \end{cases}$$

$$\sum_{i=0}^{\infty} \frac{\mathbf{B}^i t^i}{i!} = \mathbf{I} + \mathbf{Abt} + \frac{\mathbf{I}(bt)^2}{2!} + \frac{\mathbf{A}(bt)^3}{2!} + \dots =$$

$$= \begin{bmatrix} 1 + \frac{(bt)^2}{2!} + \frac{(bt)^4}{4!} + \dots & bt + \frac{(bt)^3}{3!} + \frac{(bt)^5}{5!} + \dots \\ bt + \frac{(bt)^3}{3!} + \frac{(bt)^5}{5!} + \dots & 1 + \frac{(bt)^2}{2!} + \frac{(bt)^4}{4!} + \dots \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{2}(e^{bt} + e^{-bt}) & \frac{1}{2}(e^{bt} - e^{-bt}) \\ \frac{1}{2}(e^{bt} - e^{-bt}) & \frac{1}{2}(e^{bt} + e^{-bt}) \end{bmatrix} =$$

$$= \begin{bmatrix} \cosh bt & \sinh bt \\ \sinh bt & \cosh bt \end{bmatrix}$$

Calculation of $e^{\mathbf{A}t}$ from the Cayley-Hamilton theorem

$$\det(\mathbf{A} - \lambda\mathbf{I}) = \lambda^M + \alpha_{M-1}\lambda^{M-1} + \cdots + \alpha_0 = 0$$

Matrix \mathbf{A} satisfies its characteristic equation

$$\mathbf{A}^M + \alpha_{M-1}\mathbf{A}^{M-1} + \cdots + \alpha_0\mathbf{I} = \mathbf{0}$$

Minimal polynomial

$$\mathbf{A}^J + \kappa_{M-1}\mathbf{A}^{J-1} + \cdots + \kappa_0\mathbf{I} = \mathbf{0} \quad J \leq M$$

The minimal polynomial is unique,
 λ_i are the roots of the minimal
polynomial,
but possibly of lower multiplicity.

Consider differential equation

$$\frac{d^J f(t)}{dt^J} + \kappa_{M-1} \frac{d^{J-1} f(t)}{dt^{J-1}} + \cdots + \kappa_0 f(t) = 0$$

Let a set of functions

$$f_p(t), \quad p = 0, 1, \dots, J-1$$

be the particular solutions satisfying
the initial conditions

$$f_p^{(r)}(0) = \begin{cases} 0 & \text{for } r \neq p \\ 1 & \text{for } r = p \end{cases}$$

Then

$$e^{\mathbf{A}t} = f_{J-1}(t)\mathbf{A}^{J-1} + \cdots + f_0(t)\mathbf{I}$$

An example

$$\mathbf{A} = \begin{bmatrix} 1 & 5 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} \quad \det(\mathbf{A} - \lambda \mathbf{I}) =$$

$$= \begin{vmatrix} 1 - \lambda & 5 & 0 \\ 0 & 1 - \lambda & 0 \\ 0 & 1 & 1 - \lambda \end{vmatrix} = (1 - \lambda)^3$$

$\lambda = 1$ – the triple eigenvalue

$(\mathbf{I} - \mathbf{A})^3 = \mathbf{0}$ from the Cayley-Hamilton theorem

$$\begin{bmatrix} 0 & -5 & 0 \\ 0 & 0 & 0 \\ 0 & -1 & 0 \end{bmatrix} \begin{bmatrix} 0 & -5 & 0 \\ 0 & 0 & 0 \\ 0 & -1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

The minimal polynomial ($J = 2$)
 $(\lambda - 1)^2 = \lambda^2 - 2\lambda + 1 = 0$

$$\frac{d^2 f(t)}{dt^2} - 2 \frac{df(t)}{dt} + f(t) = 0$$

solutions: $f_1(t) = e^t$ $f_2(t) = te^t$

$$g_1(t) = f(t), \quad g_2(t) = f'(t), \quad \frac{d}{dt} \begin{bmatrix} g_1 \\ g_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} g_1 \\ g_2 \end{bmatrix}$$

$$W(t) = \begin{vmatrix} f_{11}(t) & f_{21}(t) \\ f'_{11}(t) & f'_{21}(t) \end{vmatrix} = \begin{vmatrix} e^t & te^t \\ e^t & e^t + te^t \end{vmatrix} = e^{2t} > 0$$

a general solution $f(t) = \gamma_1 e^t + \gamma_2 te^t$

the particular solutions

$$f_0(0) = 1, \quad f_0^{(1)}(0) = 0: \quad f_0(t) = e^t - te^t$$

$$f_1(0) = 0, \quad f_1^{(1)}(0) = 1: \quad f_1(t) = te^t$$

$$e^{\mathbf{At}} = te^t \begin{bmatrix} 1 & 5 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} + (e^t - te^t) \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} e^t & 5te^t & 0 \\ 0 & e^t & 0 \\ 0 & te^t & e^t \end{bmatrix}$$

$$e^{\mathbf{At}} = f_{J-1}(t)\mathbf{A}^{J-1} + \cdots + f_0(t)\mathbf{I}$$

Calculation of $e^{\mathbf{At}}$ from the Laplace transform

$$e^{\mathbf{At}} = \mathcal{L}^{-1}\{(s\mathbf{I} - \mathbf{A})^{-1}\}$$

[$s \neq s_i$ - any eigenvalue]

Lemma

$$(s\mathbf{I} - \mathbf{A})^{-1} = \frac{\mathbf{B}(s)}{d(s)}$$

where

$$d(s) = \det(s\mathbf{I} - \mathbf{A}) = s^M + d_1 s^{M-1} + \dots$$

$$+ d_{M-1} s + d_M$$

$$\mathbf{B}(s) = s^{M-1} \mathbf{B}_0 + s^{M-2} \mathbf{B}_1 + \dots + s \mathbf{B}_{M-2} + \mathbf{B}_{M-1}$$

Proof: From the Cramer's rule, the (i, j) element of $(s\mathbf{I} - \mathbf{A})^{-1}$ is equal to $P_{ij}(s)/d(s)$, where $P_{ij}(s)$ is the cofactor of the (i, j) element. The cofactor is equal to $(-1)^{i+j}$ times the determinant of a $(M - 1) \times (M - 1)$ matrix.

Formulae for matrices \mathbf{B}_i

$$\mathbf{B}_0 = \mathbf{I} \quad d_1 = -\text{tr } \mathbf{A}$$

$$\mathbf{B}_1 = \mathbf{B}_0 \mathbf{A} + d_1 \mathbf{I} \quad d_2 = -\frac{1}{2} \text{tr} (\mathbf{B}_1 \mathbf{A})$$

.....

$$\mathbf{B}_k = \mathbf{B}_{k-1} \mathbf{A} + d_k \mathbf{I} \quad d_k = -\frac{1}{k} \text{tr} (\mathbf{B}_{k-1} \mathbf{A})$$

.....

$$\mathbf{B}_{M-1} = \mathbf{B}_{M-2} \mathbf{A} + d_{M-1} \mathbf{I} \quad d_{M-1} = -\frac{1}{M-1} \text{tr} (\mathbf{B}_{M-2} \mathbf{A})$$

$$d_M = -\frac{1}{M} \text{tr} (\mathbf{B}_{M-1} \mathbf{A})$$

Comment

The elements of the $(s\mathbf{I} - \mathbf{A})^{-1}$ matrix are rational functions of the variable s . It may happen that some or even all elements have common factors both in the numerator and the denominator.

Proof L.A.Zadeh, C.A.Desoer (1963) *Linear System Theory*. McGraw-Hill, New York.

An example

$$\frac{dy_1(t)}{dt} = -2y_1(t) + y_2(t)$$

$$\frac{dy_2(t)}{dt} = 2y_1(t) - 3y_2(t)$$

$$\frac{d\mathbf{y}(t)}{dt} = \begin{bmatrix} -2 & 1 \\ 2 & -3 \end{bmatrix} \mathbf{y}(t)$$

$$s\mathbf{I} - \mathbf{A} = \begin{bmatrix} s+2 & -1 \\ -2 & s+3 \end{bmatrix}$$

$$\begin{vmatrix} s+2 & -1 \\ -2 & s+3 \end{vmatrix} = s^2 + 5s + 4 = (s+1)(s+4)$$

$$(s\mathbf{I} - \mathbf{A})^{-1} = \begin{bmatrix} \frac{s+3}{(s+1)(s+4)} & \frac{1}{(s+1)(s+4)} \\ \frac{2}{(s+1)(s+4)} & \frac{s+2}{(s+1)(s+4)} \end{bmatrix}$$

$$\frac{s+3}{(s+1)(s+4)} = \frac{2}{3} \frac{1}{s+1} + \frac{1}{3} \frac{1}{s+4}$$

$$\frac{2}{(s+1)(s+4)} = \frac{2}{3} \frac{1}{s+1} - \frac{2}{3} \frac{1}{s+4}$$

$$\frac{1}{(s+1)(s+4)} = \frac{1}{3} \frac{1}{s+1} - \frac{1}{3} \frac{1}{s+4}$$

$$\frac{s+2}{(s+1)(s+4)} = \frac{1}{3} \frac{1}{s+1} + \frac{2}{3} \frac{1}{s+4}$$

$$e^{\mathbf{At}} = \mathcal{L}^{-1}\{(s\mathbf{I} - \mathbf{A})^{-1}\} =$$

$$= \frac{1}{3} \begin{bmatrix} 2e^{-t} + e^{-4t} & e^{-t} + 2e^{-4t} \\ 2e^{-t} - 2e^{-4t} & e^{-t} - e^{-4t} \end{bmatrix}$$

Equations with periodic coefficients

$$\frac{dy(t)}{dt} = \mathbf{A}(t)y(t) \quad y(0) = y_0$$

$\mathbf{A}(t)$ - piecewise continuous periodic function of t with a period $T \quad M \times M$

$$\mathbf{A}(t + T) = \mathbf{A}(t)$$

The solution

$$y(t) = \Phi(t, 0)y_0 \stackrel{\text{def}}{=} \Phi(t)y_0$$

$\Phi(t)$ - the normed fundamental matrix

$$\frac{d\Phi(t)}{dt} = \mathbf{A}(t)\Phi(t) \quad \Phi(0) = \mathbf{I}$$

$$\begin{aligned}\frac{d\Phi(t+T)}{dt} &= \mathbf{A}(t+T)\Phi(t+T) = \\ &= \mathbf{A}(t)\Phi(t+T)\end{aligned}$$

$\Phi(t+T)$ - a fundamental matrix

$\Phi(T) \ (t = 0)$ nonsingular

$$\Psi(t) \stackrel{\text{def}}{=} \Phi(t+T)\Phi^{-1}(T)$$

$$\frac{d\Psi(t)}{dt} = \mathbf{A}(t)\Psi(t) \quad \Psi(0) = \mathbf{I}$$

$\Psi(t)$ - the normed fundamental matrix

$$\Psi(t) = \Phi(t)$$

$$\Phi(t+T) = \Phi(t)\Phi(T)$$

$$\Phi(t+nT) = \Phi(t)\Phi^n(T)$$

$$y(t+nT) = \Phi(t)\Phi^n(T)y_0$$

$$y(nT) = \Phi^n(T)y_0$$

Equations with periodic coefficients

consider an equation

$$\frac{d\mathbf{x}(t)}{dt} = \mathbf{B}\mathbf{x}(t) \quad \mathbf{x}(0) = \mathbf{y}_0$$

\mathbf{B} – a constant matrix

$$\mathbf{x}(t) = e^{\mathbf{B}t} \mathbf{y}_0$$

require $\mathbf{y}(nT) = \mathbf{x}(nT)$

$$\Phi^n(T) \mathbf{y}_0 = (e^{\mathbf{B}T})^n \mathbf{y}_0$$

$\mathbf{C} = [\mathbf{y}_0^1, \mathbf{y}_0^2, \dots, \mathbf{y}_0^M]$ - chosen to be nonsingular

$$\Phi^n(T) \mathbf{C} = (e^{\mathbf{B}T})^n \mathbf{C} \quad \times \mathbf{C}^{-1}$$

$$n = 1$$

$$\Phi(T) = e^{\mathbf{B}T}$$

$$\Phi(t) = \Phi(t) e^{-\mathbf{B}t} e^{\mathbf{B}t} \stackrel{\text{def}}{=} \mathbf{P}(t) e^{\mathbf{B}t}$$

A fundamental matrix in the Floquet normal form

$$\boxed{\mathbf{y}(t) = \mathbf{P}(t) e^{\mathbf{B}t} \mathbf{y}_0} \quad \mathbf{P}(0) = \Phi(0) = \mathbf{I}$$

The Floquet solution

$\mathbf{P}(t)$ periodic, as

$$\mathbf{P}(t+T) = \Phi(t+T) e^{-\mathbf{B}(t+T)} =$$

$$\Phi(t) \Phi(T) e^{-\mathbf{B}t} e^{-\mathbf{B}T} =$$

$$\Phi(t) e^{\mathbf{B}T} e^{-\mathbf{B}T} e^{-\mathbf{B}t} =$$

$$= \Phi(t) e^{-\mathbf{B}t} = \mathbf{P}(t)$$

(exponents
commute)

$\mathbf{P}(t)$ continuous and periodic, thus bounded

Thank you!